

# **Source-to-Outcome Modeling**

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# Information in this presentation addresses:

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- Chapters in the report
  - 5. Dietary exposure modeling
  - 6. Linking exposure models
  - 7. Longitudinal exposures
  - 8. Using model-to-measurement comparisons to evaluate model predictions
- Charge questions
  - 2. Longitudinal dietary exposure assessment
  - 4. Comparison of model predictions with human monitoring data

# **Modeling Oral Doses from Dietary Exposures**

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# Exposure modeling

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- Single day dietary exposure modeling has been performed since the late 1980s using the same basic approach
- Multiple models have been developed and evaluated by the SAP
- Conclusion from multiple model reviews
  - While models differ in certain details their results are similar for single day exposures
- Longitudinal (multiple consecutive days for one individual) is a greater challenge
  - Limited empirical data on longitudinal exposures
  - Multiple methods proposed to simulate longitudinal exposures

# Exposure modeling in this project

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- The PBPK/PD model is not directly linked to the exposure model - exposure data from any dietary model can be used
- Models used in this project:
  - CARES 3.0 and LifeLine 4.1
  - DEEM™ 2.16 used in model comparison
- Residue data
  - Reflect recent dietary intakes (2006-2008)
  - Developed following existing EPA regulatory policies and public examples
  - Details given in Attachment C “2010 Update of the Acute Dietary Risk Assessment for Chlorpyrifos”

# Results

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- Models give similar estimates of daily doses
  - Within a factor of 2-3 at the 99.9<sup>th</sup> percentiles
- Doses
  - Highest in 3 year olds
  - 10 - 20% lower in infants
  - 50% lower in adults
- Consistent with age-related differences in exposures to residues of other pesticides

# **Linking Dietary Exposure and PBPK/PD Models**

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# Defining the individual

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- The exposure modeling precedes the PBPK/PD modeling
- The PBPK model assumptions on physiology and metabolism need to be consistent with dose estimates and an oral pathway of exposure
- Exposure models define age, gender, and bodyweight
- Output of model is therefore:
  - Age, gender, and bodyweight
  - Longitudinal dose history



# Moving data between the models

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- Data from dietary model is converted to an Excel file;

<b>Age (y)</b>	<b>Sex 0=M, 1=F</b>	<b>Weight (kg)</b>	<b>Dose Day 1 (mg/kg)</b>	<b>Dose Day 2 (mg/kg)</b>	<b>Dose Day 3 (mg/kg)</b>	<b>Dose Day 4 (mg/kg)</b>	<b>Dose Day 5 (mg/kg)</b>
0.5	1	6.8	3.97E-06	4.39E-07	1.12E-05	1.04E-05	1.65E-06
0.5	1	6.4	3.31E-06	9.61E-07	1.12E-05	6.78E-05	1.65E-06

- The PBPK/PD model reads this file as an input

# Models view doses differently

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- Exposure models define dose as:
  - Mass of residue in daily diet on a bodyweight basis (mg/kg/d)
  - Timing of dose within a 24 hour period typically not considered
- PBPK/PD model defines dose as:
  - Mass entering into the intestine compartment
  - Rate is not constant over time
- Mass entering intestine is driven by:
  - Timing of meals and fraction of dose in each meal
  - Rate of stomach emptying

# Converting dietary doses to intestine loading rates

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- Dietary doses:
  - Occur during multiple eating events in a day; but
  - Top 2% of population receive 75% of daily dose from one food
  - Assumed total daily dose occurs at a single meal
  - Assumption is conservative (increases estimate of peak levels in blood)
- Stomach emptying rates:
  - Time to empty first half of a meal reported as 30-90 minutes
  - A value of 90 minutes is used in the model
  - Model not sensitive for rate of transfer into the intestine

# **Modeling Longitudinal Exposures**

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# Challenge of modeling longitudinal exposures

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- Current dietary surveys do not collect data on an individual's dietary intakes on consecutive days
- This requires exposure modelers to simulate longitudinal dietary exposures
  - Multiple approaches have been proposed
- This introduces uncertainty in risk predictions for exposures longer than one day

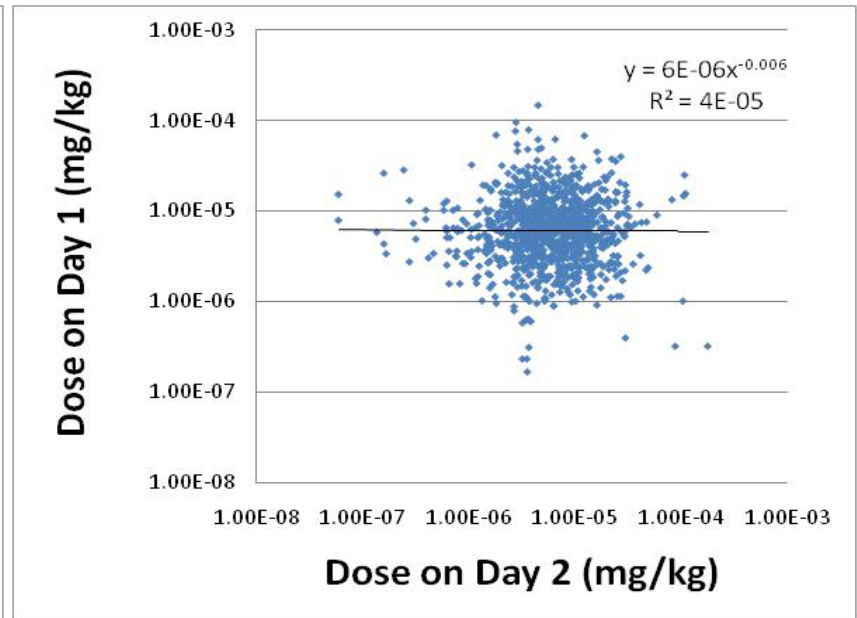
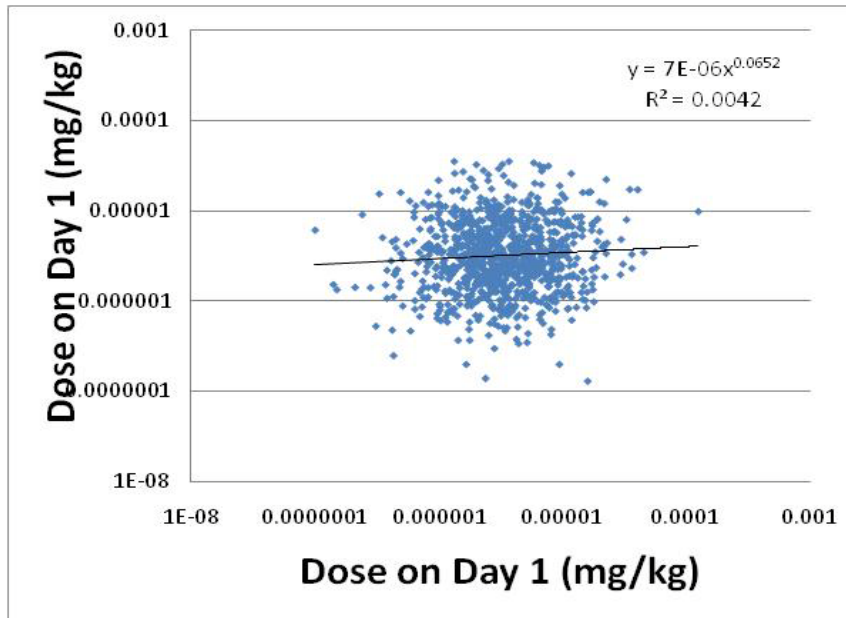
# Approach used

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1. Continuous exposures to constant doses
2. Different approaches for simulating longitudinal exposures
  - LifeLine seasonal, LifeLine daily, and CARES
3. Investigate impact of longitudinal exposures on adults with doses at or above 99.9<sup>th</sup> percentile dose

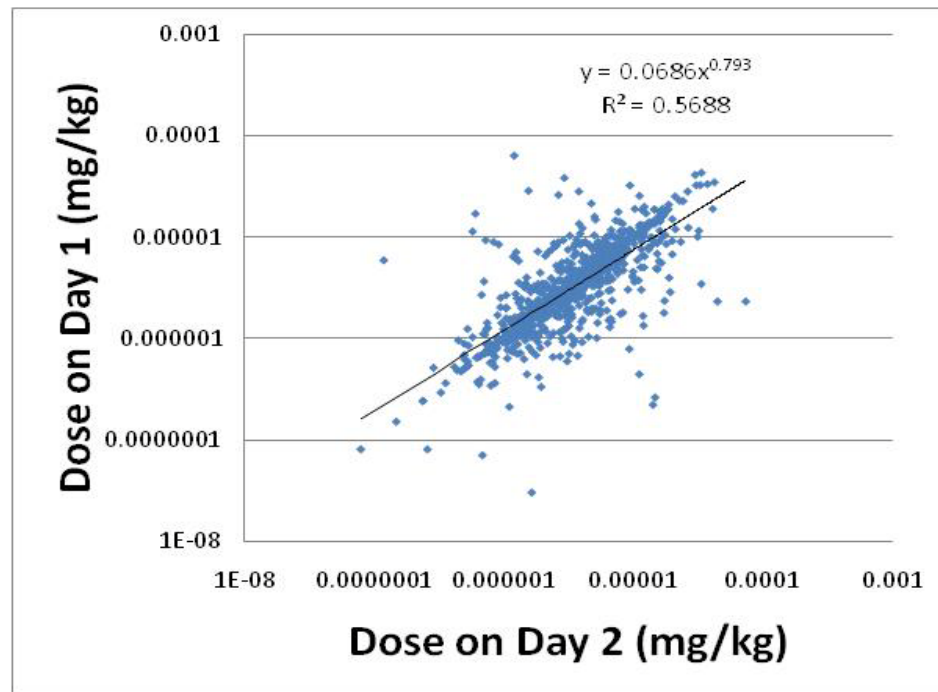
# LifeLine random day and CARES

No correlation between doses on consecutive days.



# The LifeLine seasonal

## Moderate correlation





# Investigating impacts of longitudinal exposures: bounding the issue

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All sets of data can be placed on a common scale



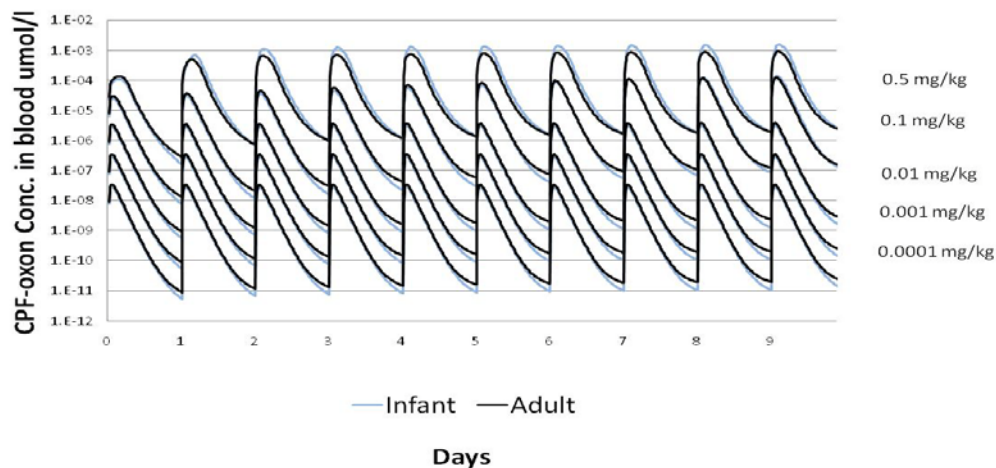
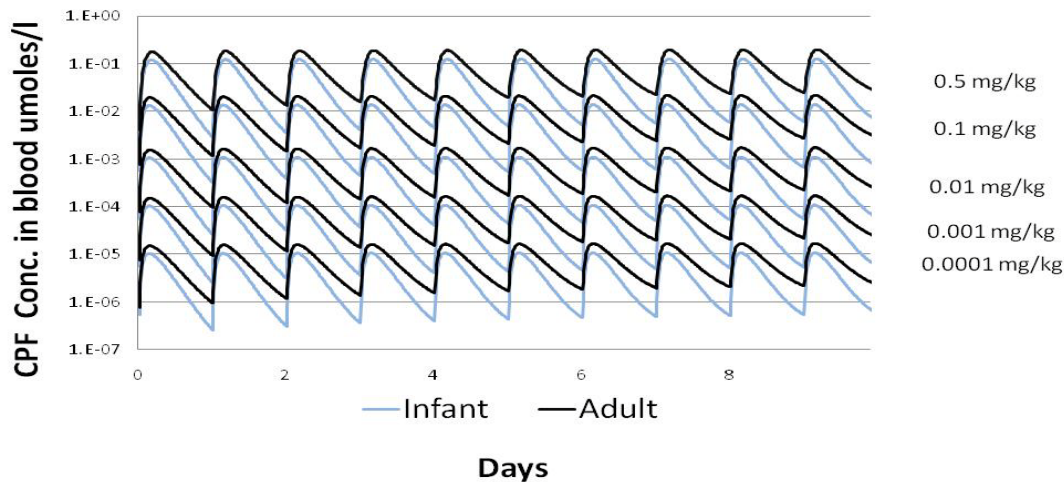
# Impact of a constant dose

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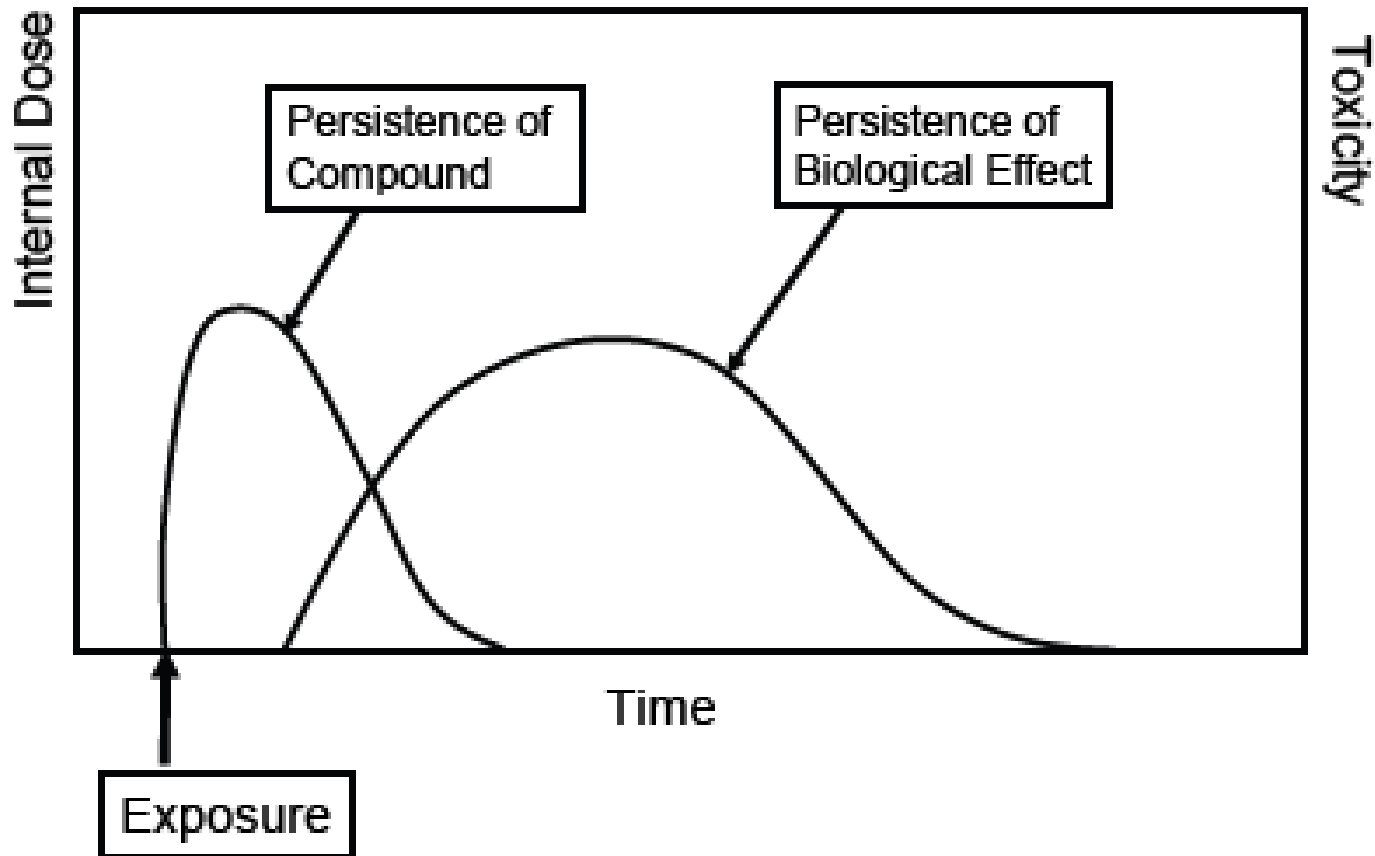
- Continuous exposures to constant dose were modeled for typical adult and infant over 30 days
- Five doses (0.5- 0.0001 mg/kg)
  - Doses that cause significant AChE inhibition (0.5 mg/kg)
  - Doses that occur on one day to the top 1% of 3 year olds (0.0001 mg/kg)
- Determined
  - Build up of CPF and CPF-oxon
  - Increasing inhibition of AChE

# CPF and CPF-oxon levels in blood

- Levels of CPF increase slightly
  - Approximately 16% increase in peak level on day 30 vs. day 1
  - Increase not dose dependent
- Levels of CPF-oxon increase in a dose dependent manor:
  - 18 fold increase at 0.5 mg/kg
  - 1% increase at 0.0001 mg/kg

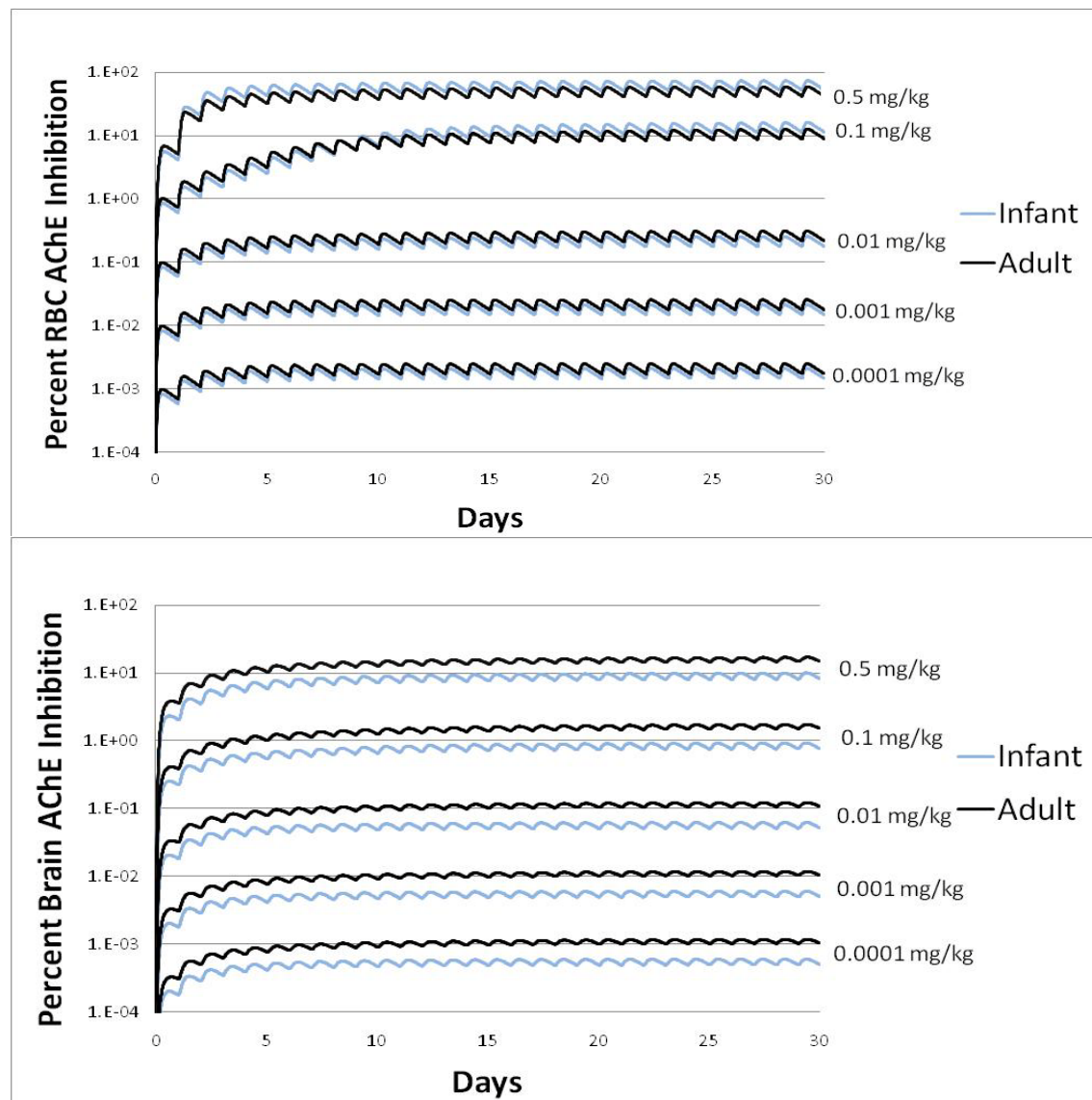


# Persistence



# RBC and Brain AChE Inhibition

- RBC AChE inhibition increases in a dose and age dependent manner:
  - 0.1 mg/kg adults had a 20 fold increase and infants a 12 fold increase
  - At 0.0001 mg/kg there was a 2.5 fold increase for both ages
- Brain AChE inhibition increases similar for adults and infants
  - Increases ranged from 4 fold at high dose to 3 fold at low dose



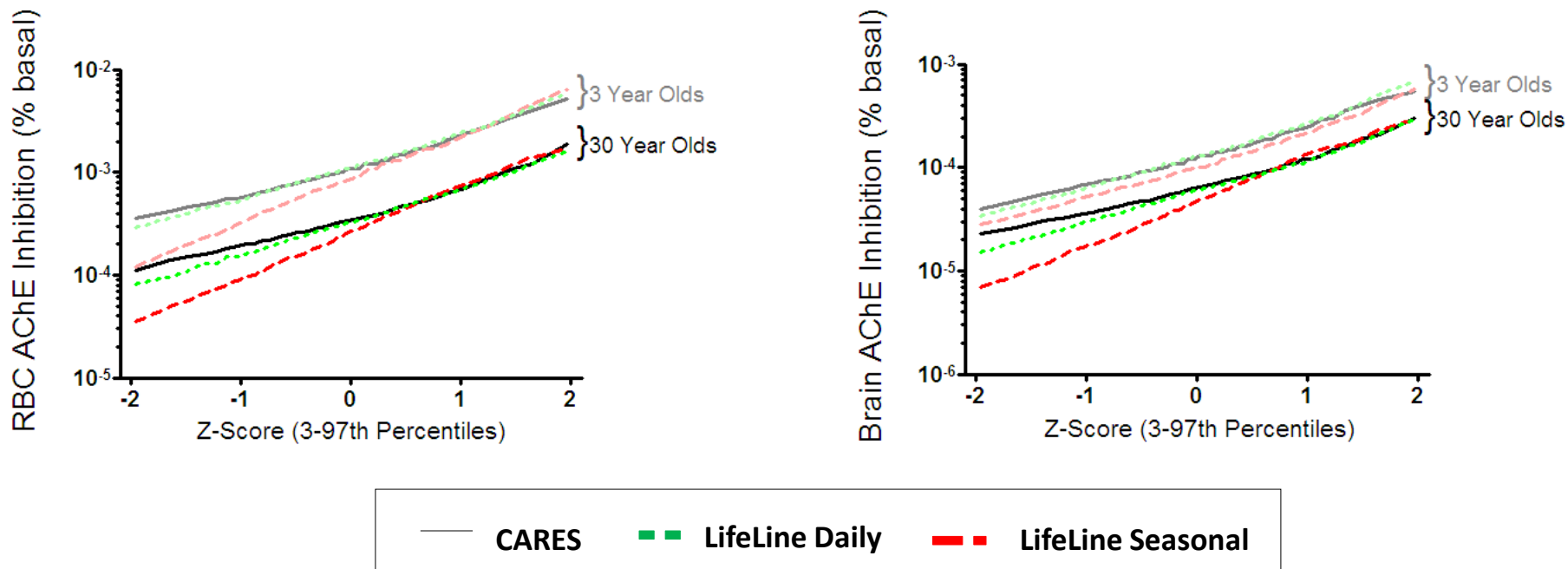
# Predictions of the three longitudinal approaches for CPF and CPF-oxon

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- All three approaches give similar results
- Consistent with the finding that CPF and CPF-oxon formed on one day do not carry over to the following days
  - If there is no “carry over” then differences in assumptions about exposures on prior days should have no impact

# Predictions of the three approaches for AChE inhibition

Approaches differ at low exposures but agree for high exposures (top one third)



# Why are responses in high-dose individuals independent of modeling assumptions?

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- Asymmetrical effects of high and low doses on subsequent days
  - A high dose (1000) will affect the next day even if only 10% carries over if the next day's dose is small (10)
  - $1000 * 0.1 + 10 = 110$  versus 10 (*a 11 fold change*)
  - In contrast, if the days are reversed
  - $10 * 0.1 + 1000 = 1001$  versus 1000 (*a 0.1% change*)
- Therefore individuals receiving smaller dose are more sensitive to doses on prior days than individuals receiving high doses
  - More affected by differences in the longitudinal models



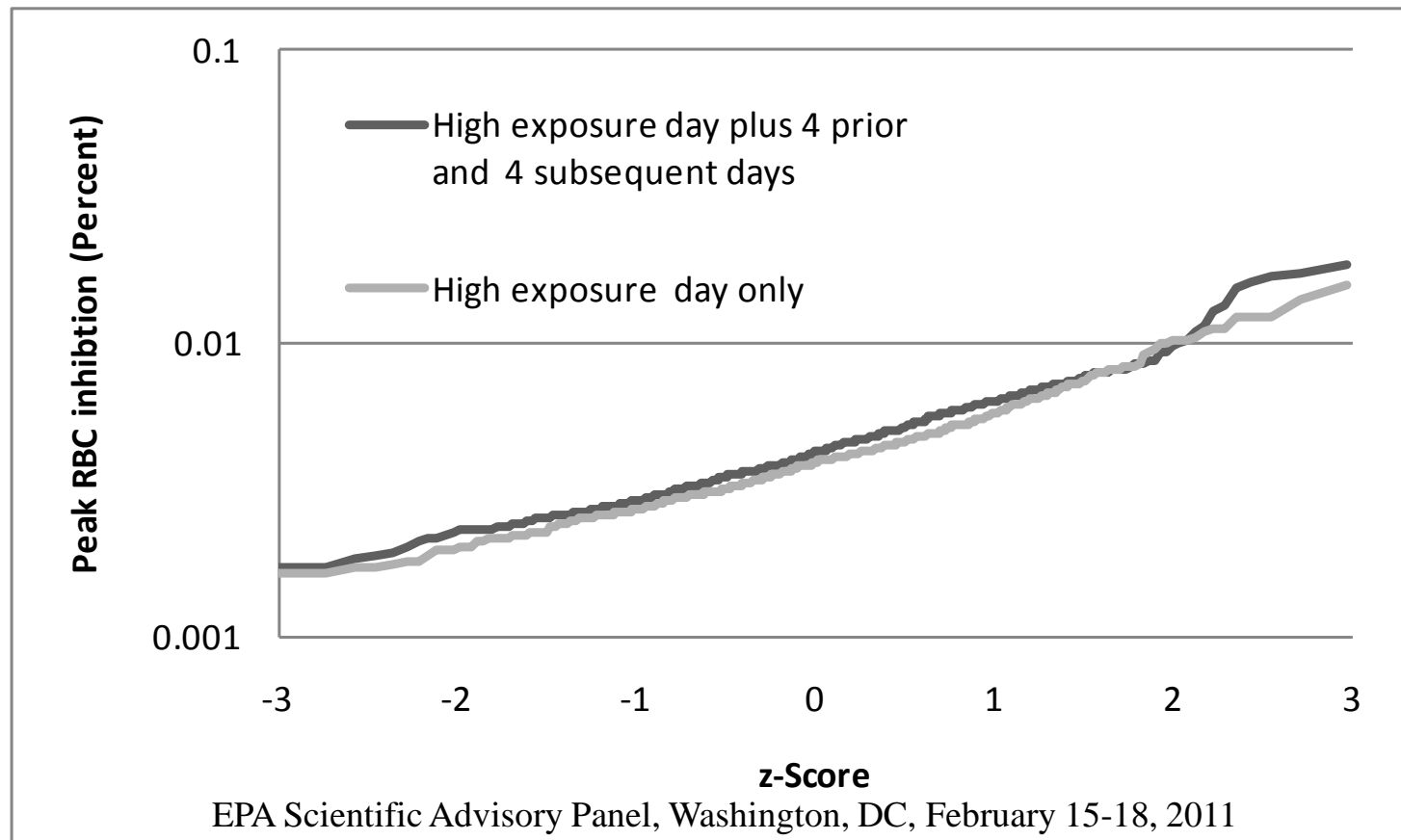
# Impact of longitudinal exposures on AChE inhibition in highly-exposed individuals

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- Procedure:
  - Dietary dose histories (365 days) were created for a large number of adults using CARES
  - Individuals were identified who had a “high” daily exposure (in the top 0.1% of the daily exposures received by adults) on one of the 365 days
  - The doses on the four days prior, and subsequent, to the high exposure day were identified
    - » Resulting in nine days of exposure with the high daily exposure occurring on day five
- The Variation model was used to determine impact of dietary exposures on peak RBC and brain AChE inhibition for:
  - The peak day alone; and
  - The nine days of exposures

# Impact of longitudinal exposures on high exposure days

Prior and subsequent exposures increased RBC AChE inhibition by an average of 10%



# Findings on Longitudinal Exposures

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- The findings of the various analyses lead to the same conclusions on the impacts of longitudinal dietary exposures
  - CPF and CPF-oxon do not accumulate over time at dietary doses
  - AChE inhibition does increase but the impact is minimal (10%) at the 99.9<sup>th</sup> percentile
- AChE effects from dietary exposures
  - Highly exposed individuals can be predicted based on single daily doses
  - Typical- and low-exposed individuals require consideration of longitudinal exposures

# **Evaluating the Source-to-Outcome Model by Using Human Monitoring Data**

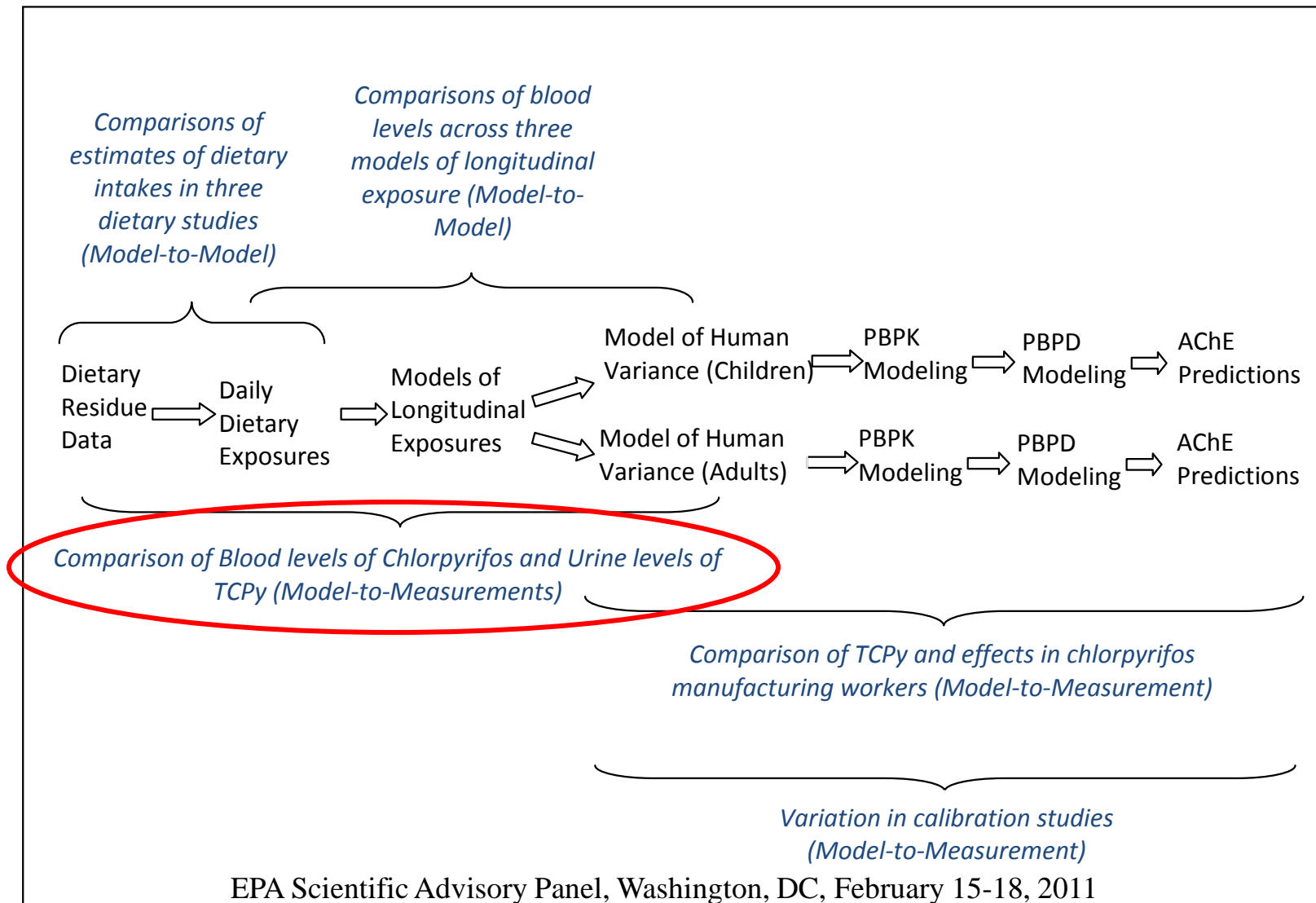
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# Model-to-Measurement

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- Source-to-outcome models and measured data
  - Recognizing the challenges of model evaluation
- Key strategy in model-to-measurement comparisons
- Comparisons to blood levels of CPF
- Comparisons to urinary measurements of TCPy

# No one set of monitoring data can evaluate the entire source-to-outcome model



# Strategy

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- Measured data is produced by surveys with “characteristics” that affect their findings
  - Ages surveyed
  - Time since last exposure
  - Pregnancy
- To facilitate model-to-measurement the source-to-outcome model needs to predict results as the survey would measure them
  - Mimic survey characteristics

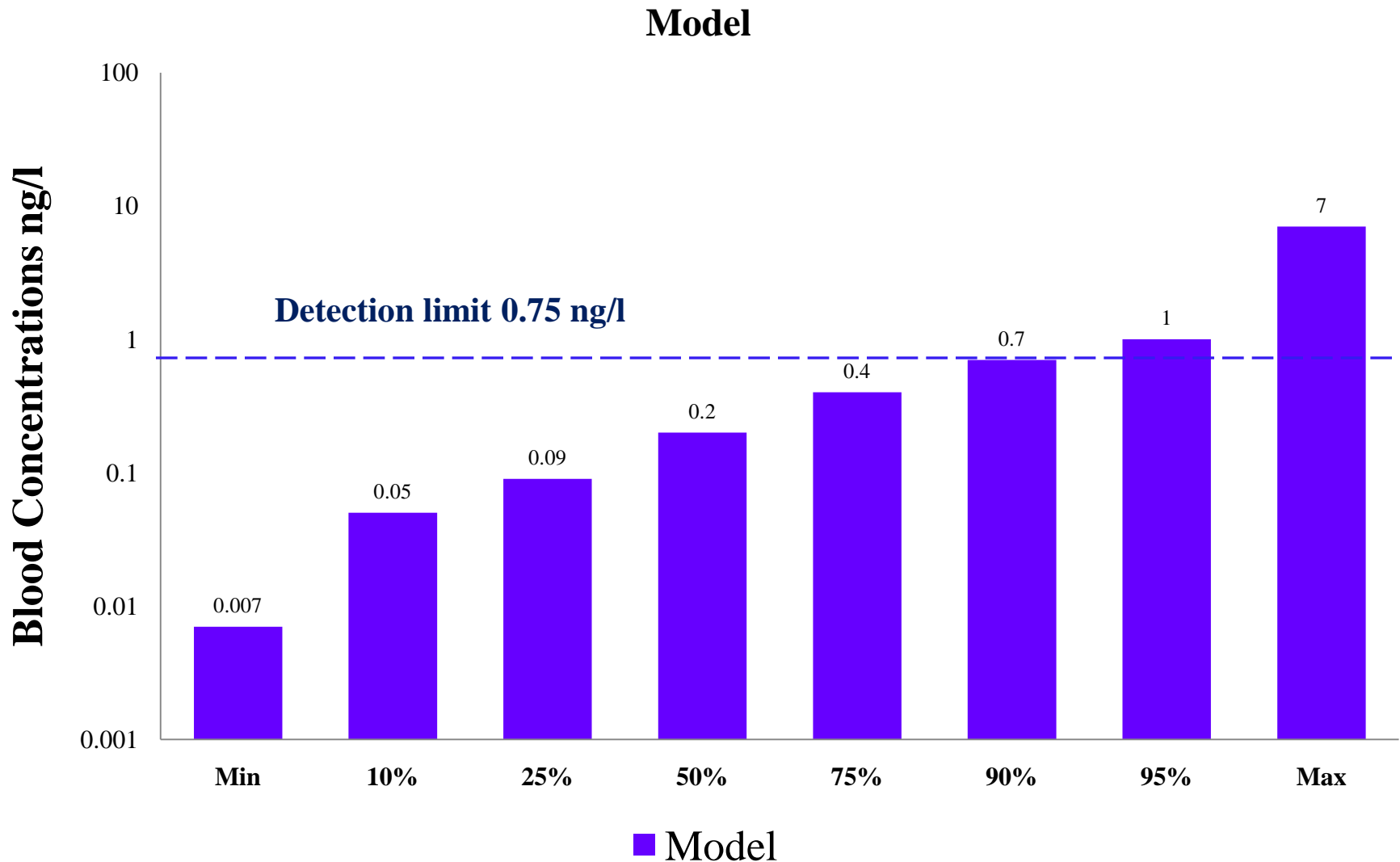
# Comparison to survey data on blood levels of CPF

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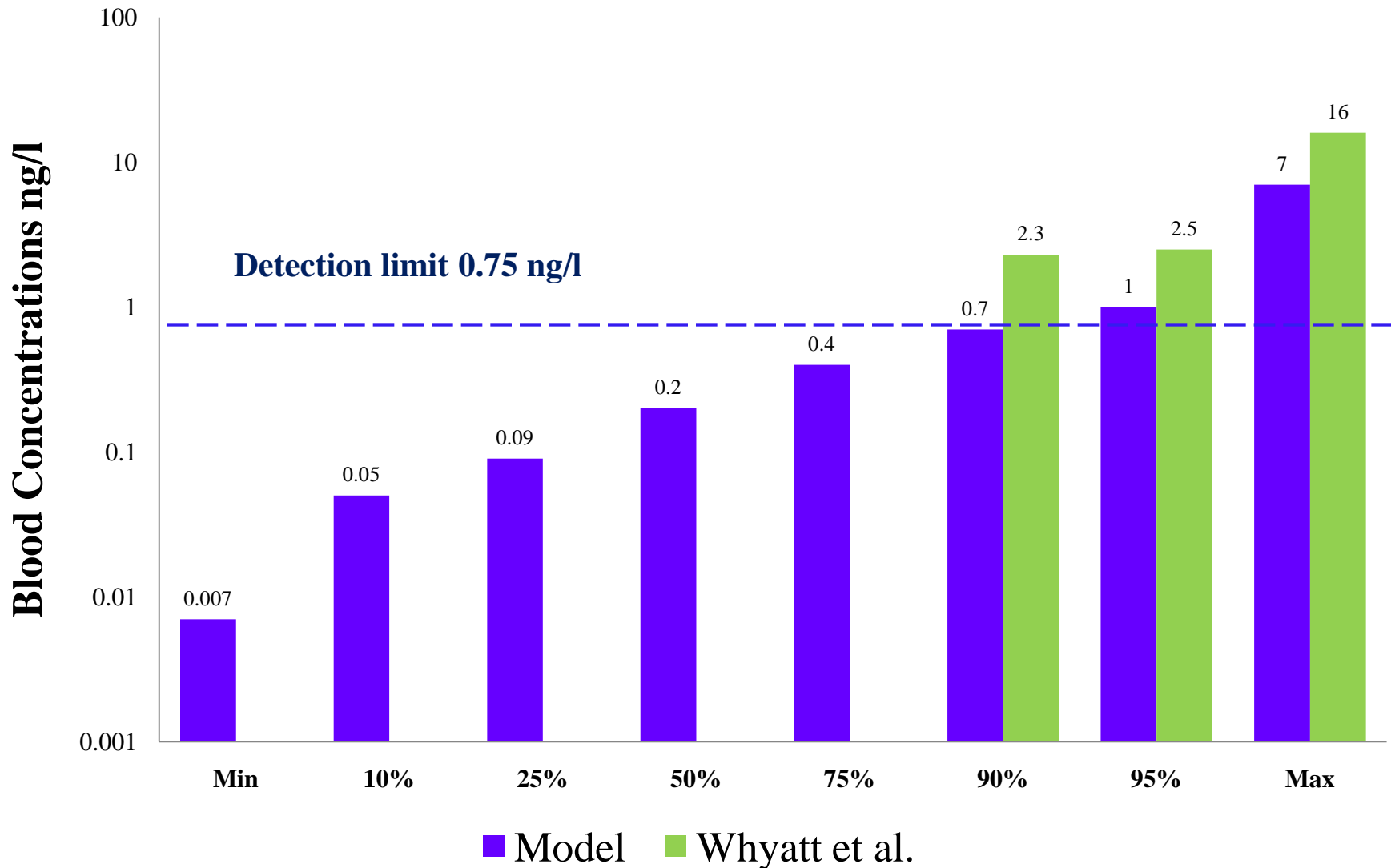
- Current levels are available from two surveys Whyatt et al. and Barr et al.
  - Data collected from mothers at the time of delivery
  - Exposures believed to be limited to dietary exposures
  - The time since last meal is not reported
- Compare to model results
  - Variation across adults
  - Levels in blood of adults at any time between 0 and 24 hours following a dietary exposure
  - Adjusted to reflect pregnancy



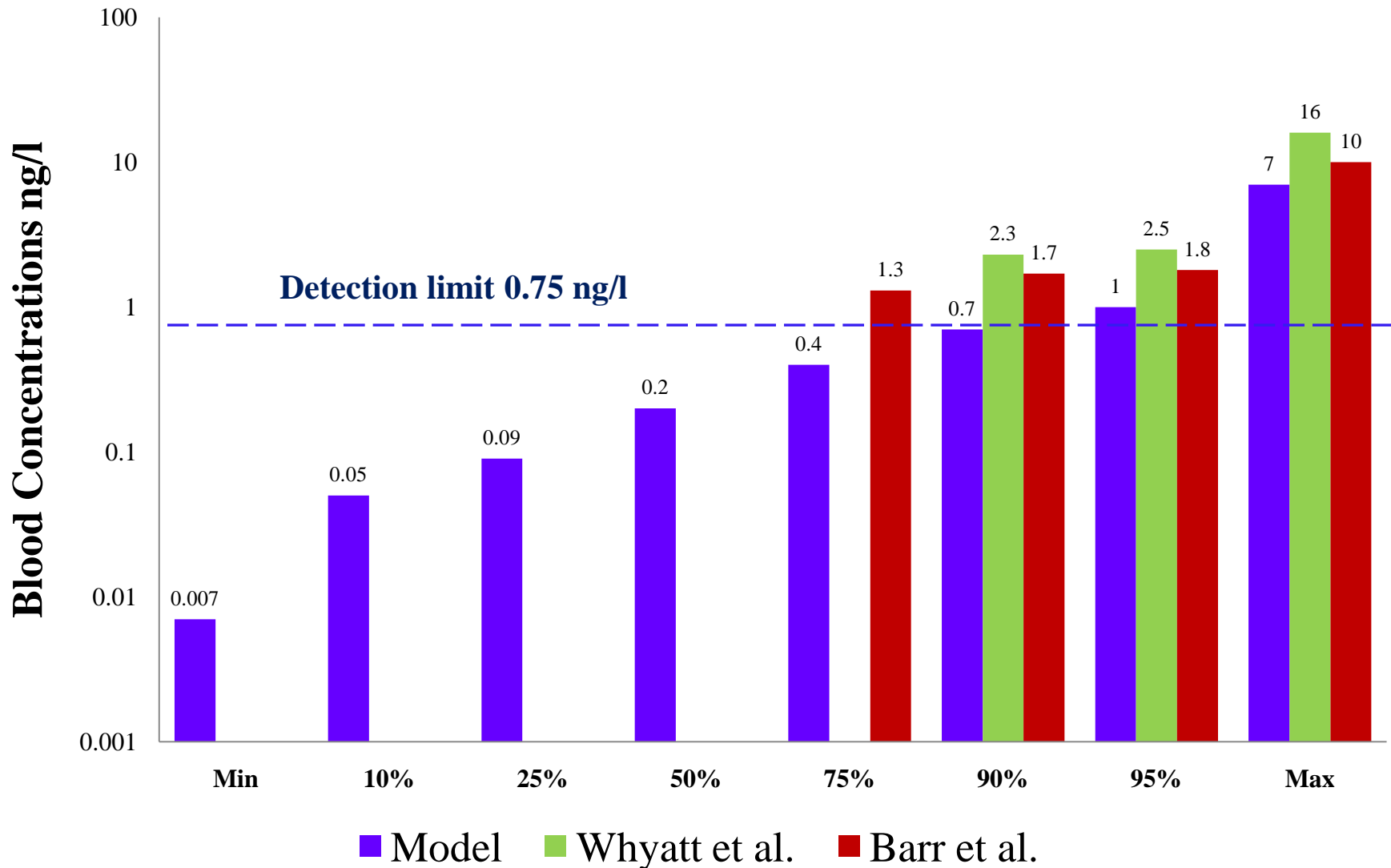
# Model-to-Measurement Results



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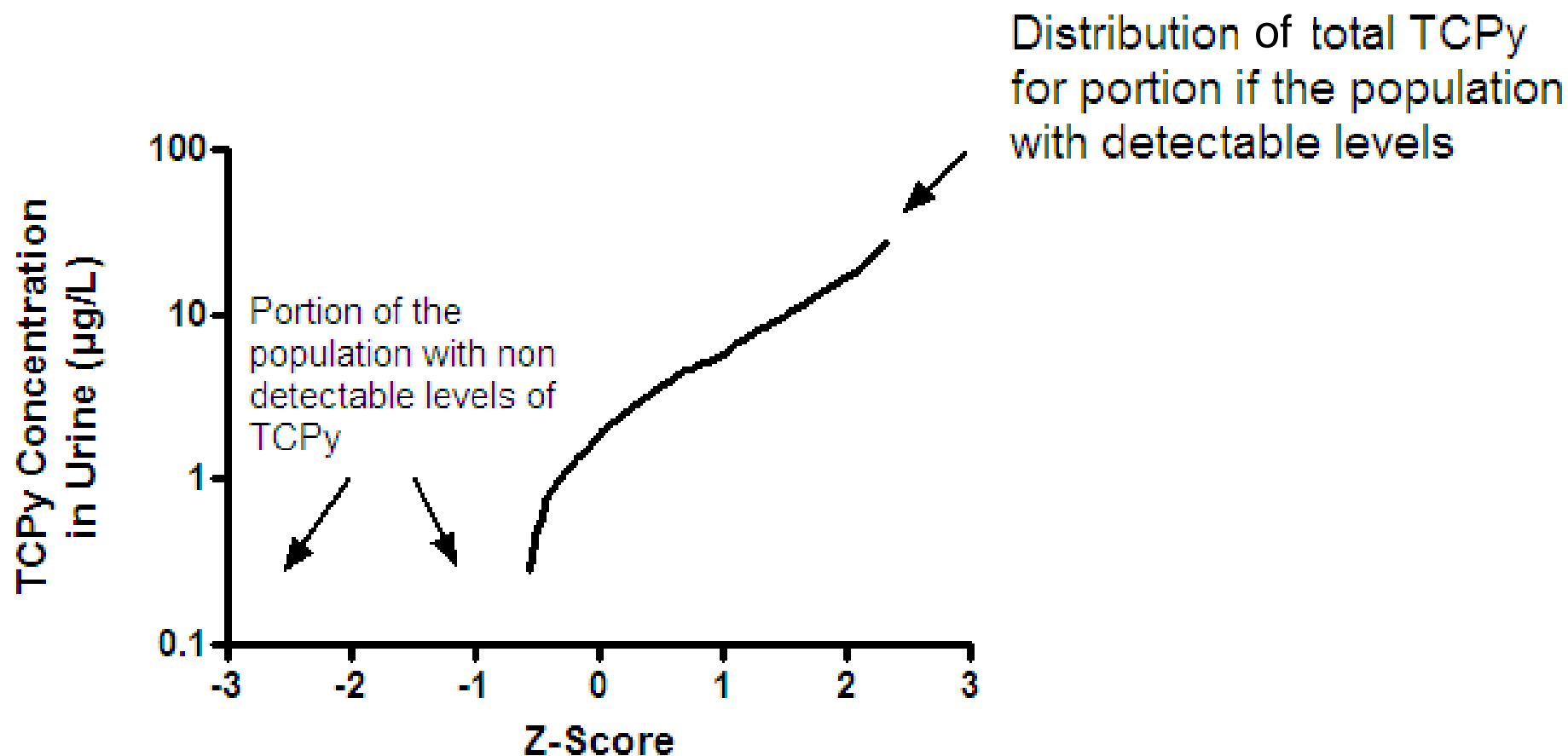


# Comparison to TCPy data

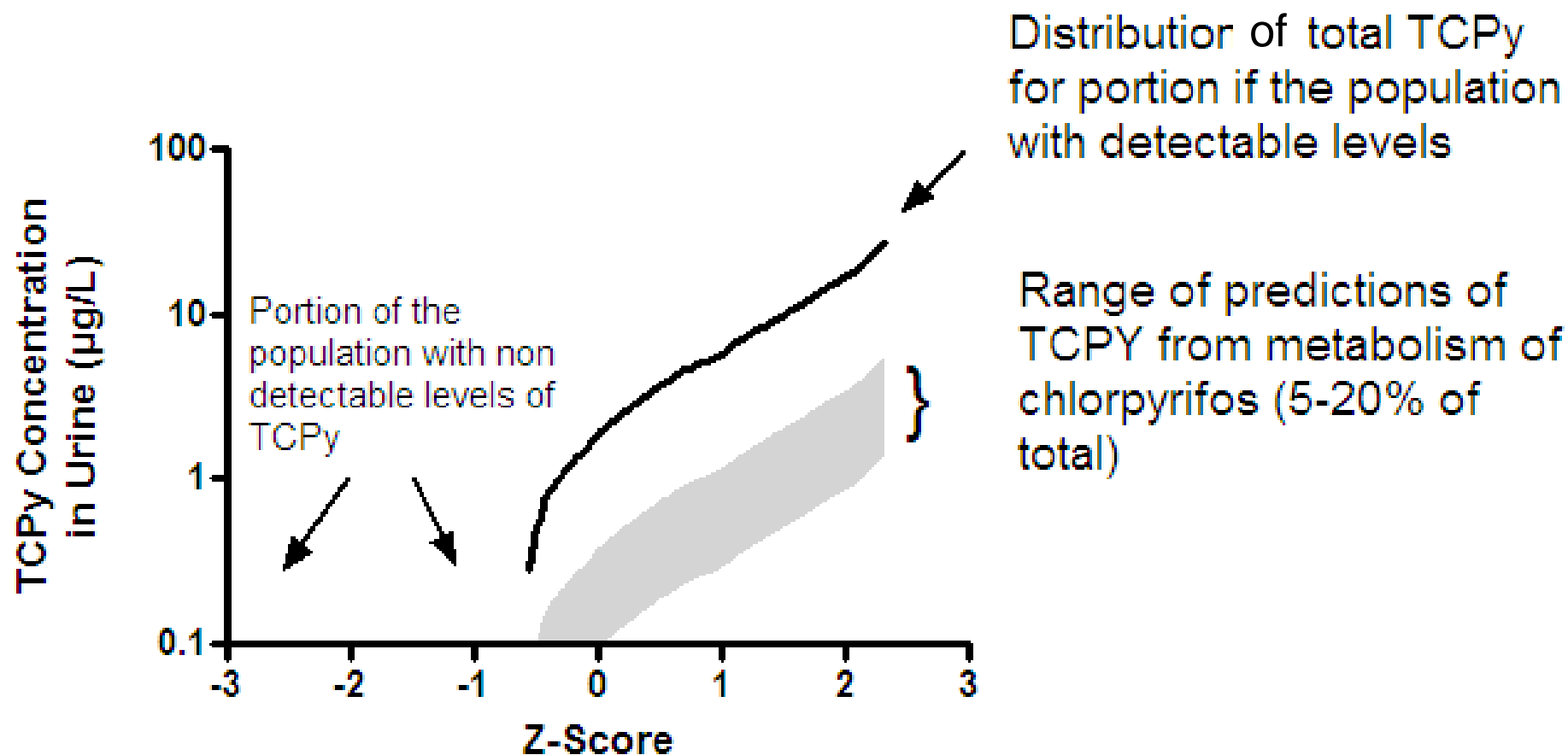
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- NHANES data:
  - Data collected in 2002 limited to 12yr and older
  - Spot urine sample
- NHANES data on TCPy reflects:
  - Dietary intakes of TCPy, CPF, and methyl-CPF
  - TCPy produced by metabolism of CPF residues in diet is believed to be between 5 and 20% of total TCPy

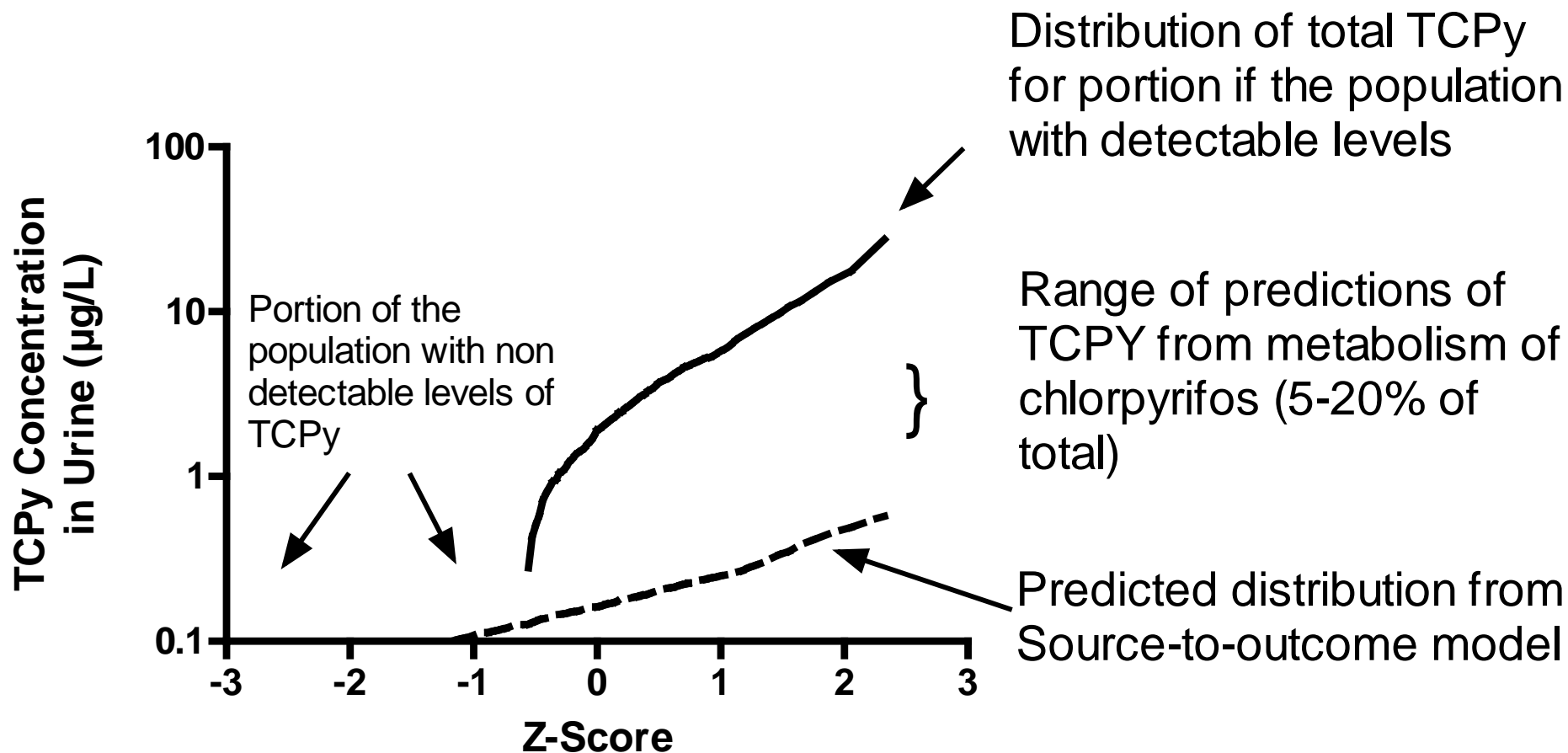
# Comparison to NHANES data



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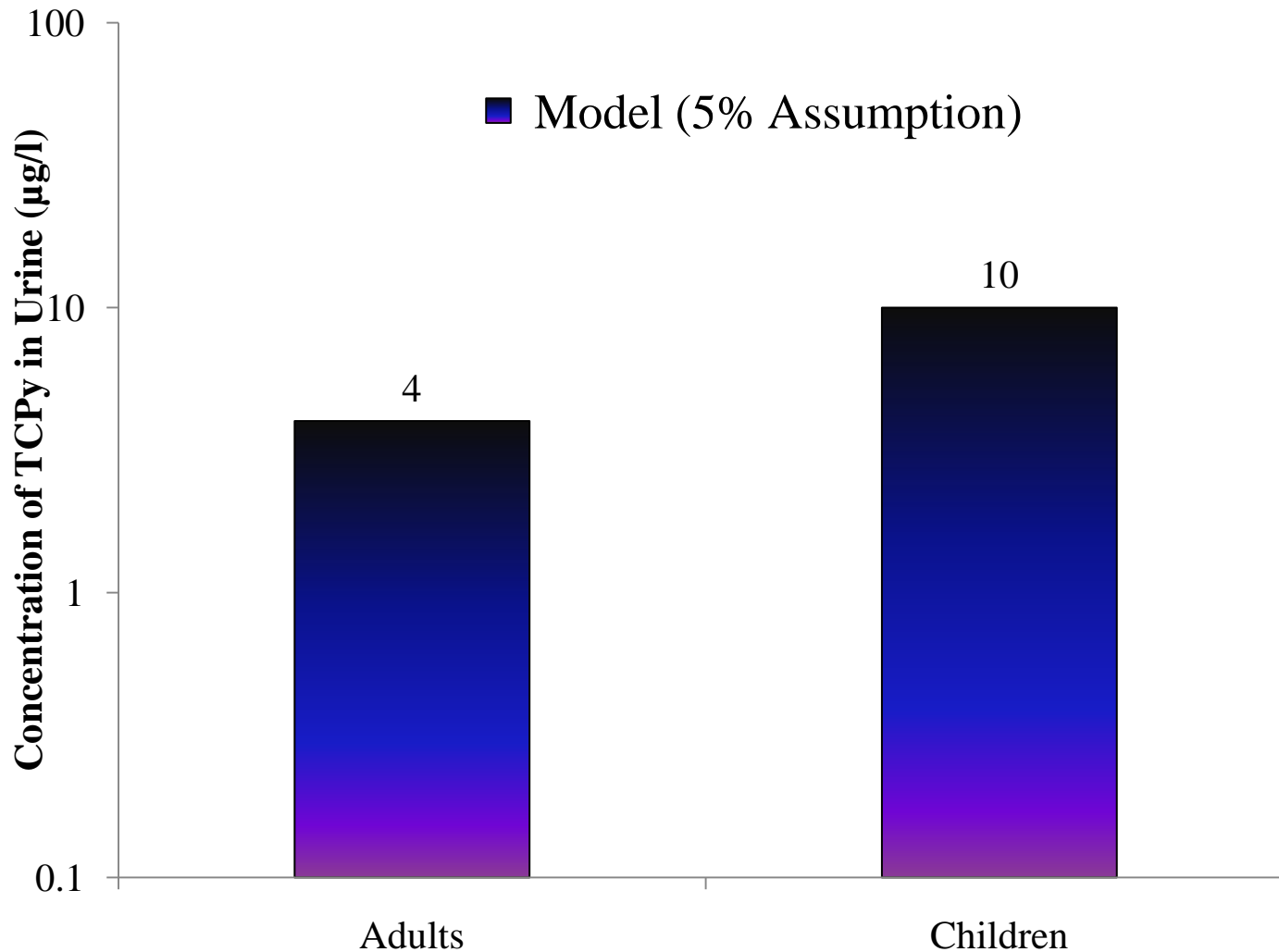


# Comparison to NHANES data



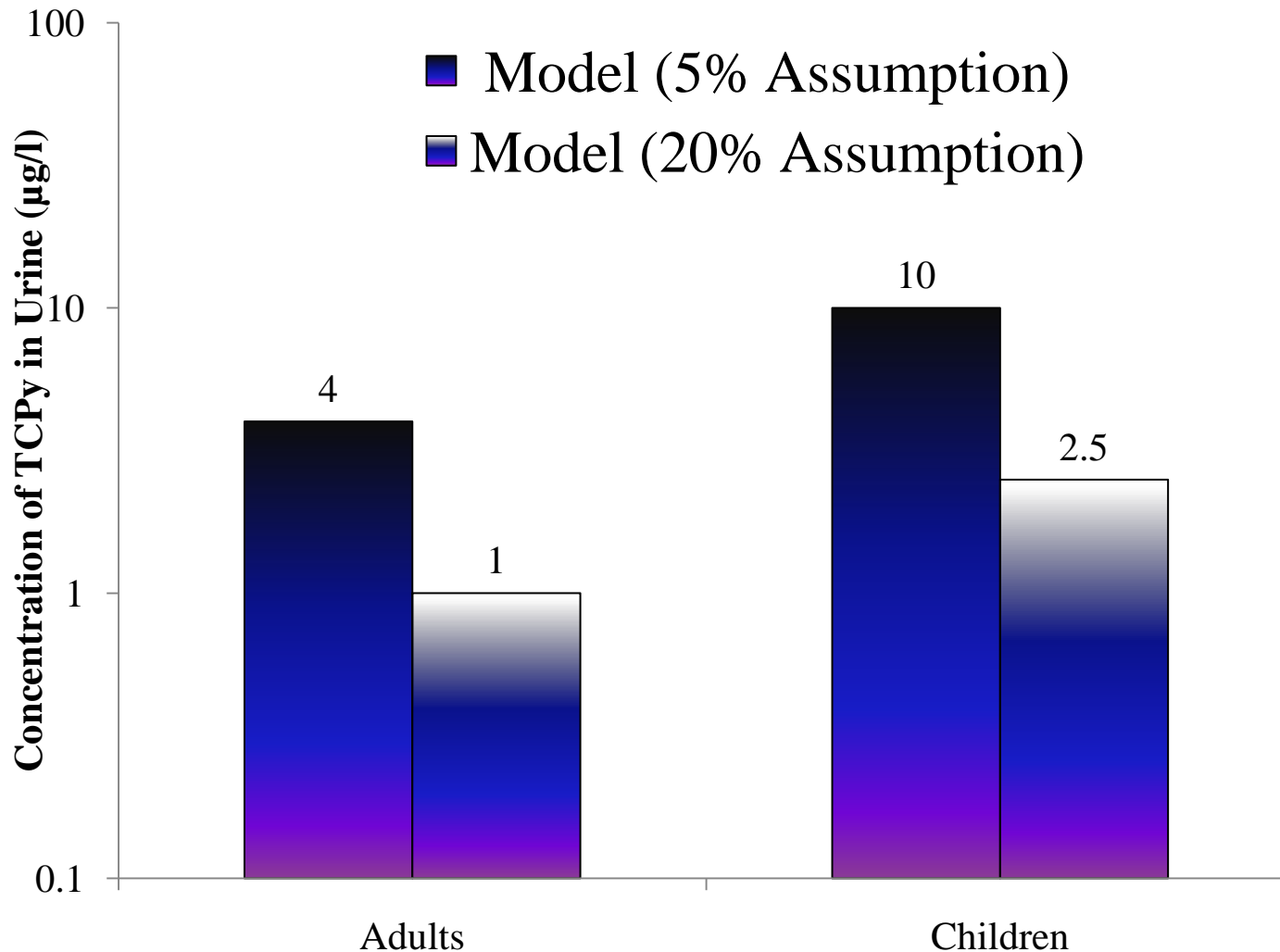
# Comparison of observed levels of TCPy in urine and model predictions

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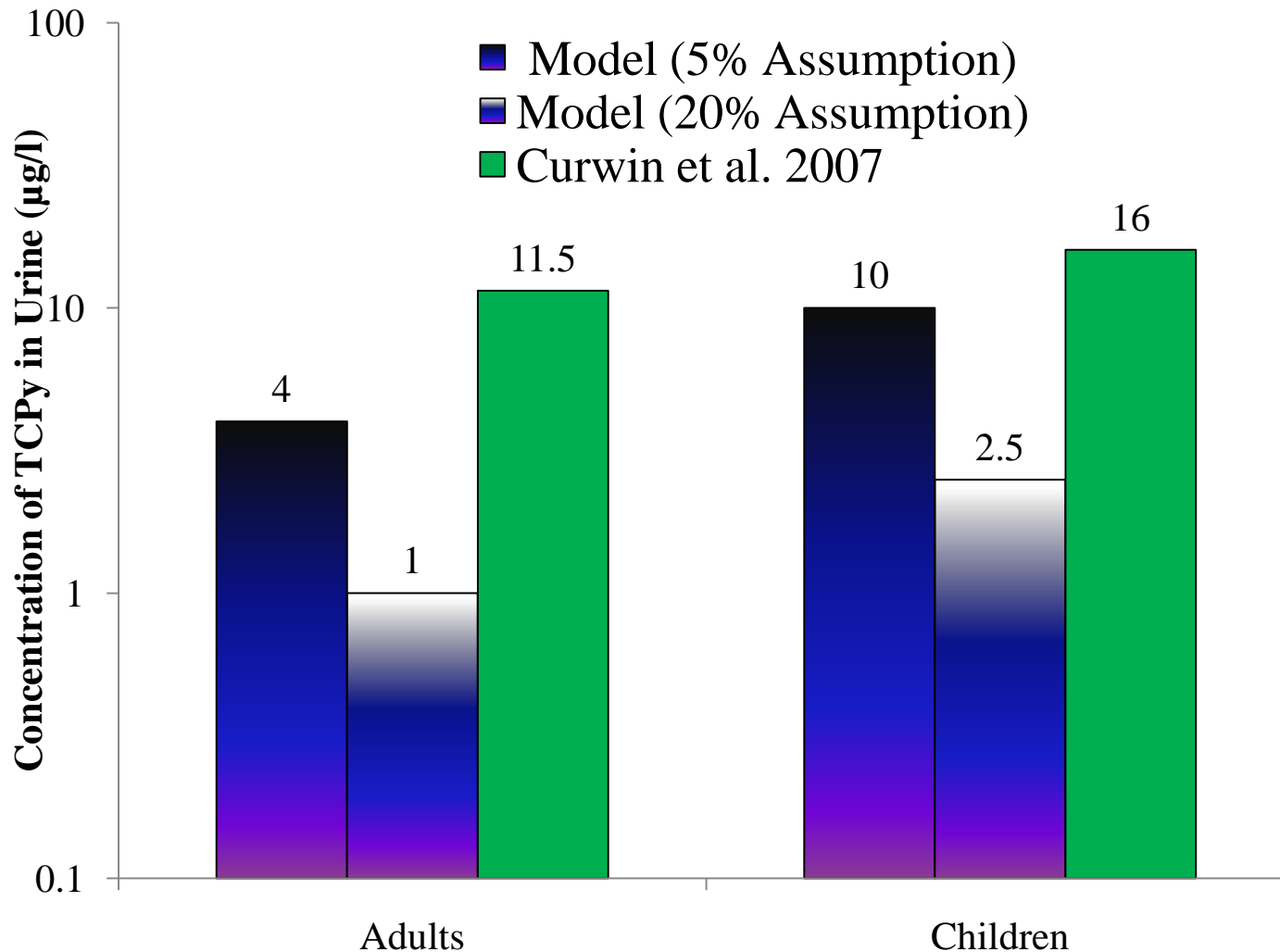




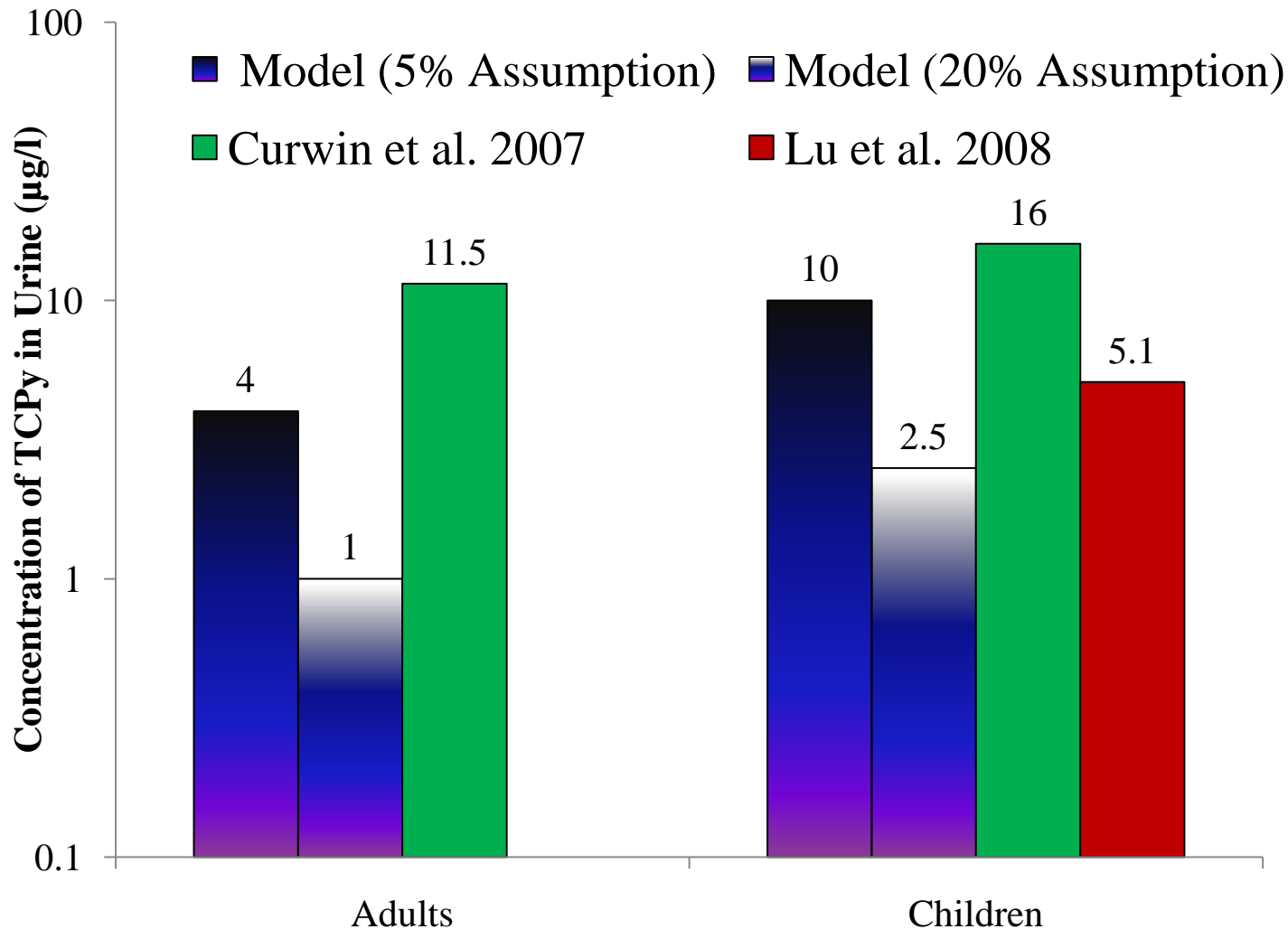
# Comparison of observed levels of TCPy in urine and model predictions



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# Conclusions

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- The portions of source-to-outcome model have been evaluated by comparison to human volunteer studies
  - Measure of central tendency for groups of individuals (LifeStage model)
  - Measure variation across individuals (Variation model)
  - Variation model over predicts response
- The predictions of the exposure and PBPK portions of the model matched the monitoring data